

# CSE 559A: Computer Vision



[credit: danjodon.deviantart.com]

Fall 2017: T-R: 11:30-1pm @ Lopata 101

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<http://www.cse.wustl.edu/~ayan/courses/cse559a/>

Sep 19, 2017

# ADMINISTRIVIA

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- Recitation will be this Friday (9/22) in J309.
  - Will go over topics relevant to Pset.
- Yesterday office hours canceled last minute
  - Q: Useful to have another set this week ?

# GENERAL

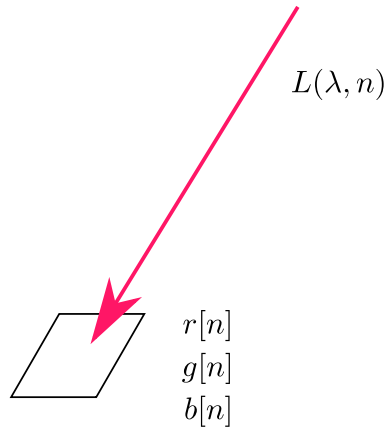
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## Suggestions about Pset

- Look at the support code !
- Look at math / coding resources on course website
- Problem 3:
  - Direction of  $\theta$  depends on convention. NMS fig in lecture slides is just an example.
  - Any convention is correct, as long as it is internally consistent.
  - LOOK at the results and see if your code is doing the right thing !
  - The idea is that you should 'thin' edges to select maxima.
- Problem 4:
  - Support code calls bilateral filter multiple times deliberately (to encourage you to write efficient code!)
  - If you want to debug, temporary comment out later calls (or just press ctrl+c after first file saved).

# COLOR

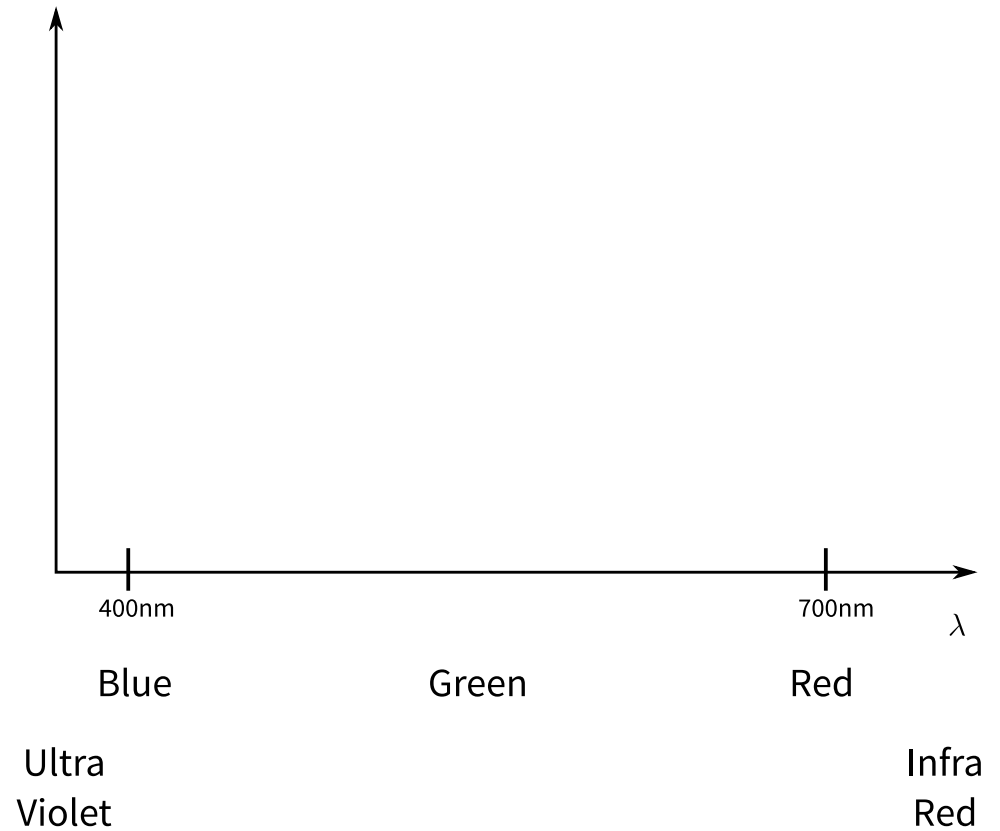
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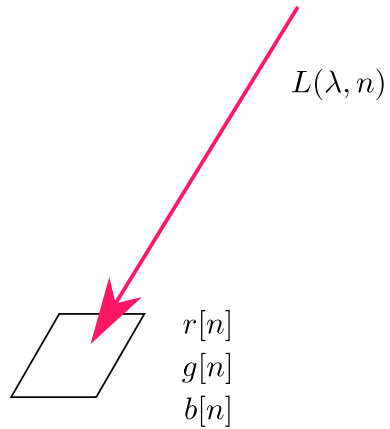
$$r[n] = \int L(\lambda, n) \Pi_r(\lambda) d\lambda$$

$$g[n] = \int L(\lambda, n) \Pi_g(\lambda) d\lambda$$

$$b[n] = \int L(\lambda, n) \Pi_b(\lambda) d\lambda$$



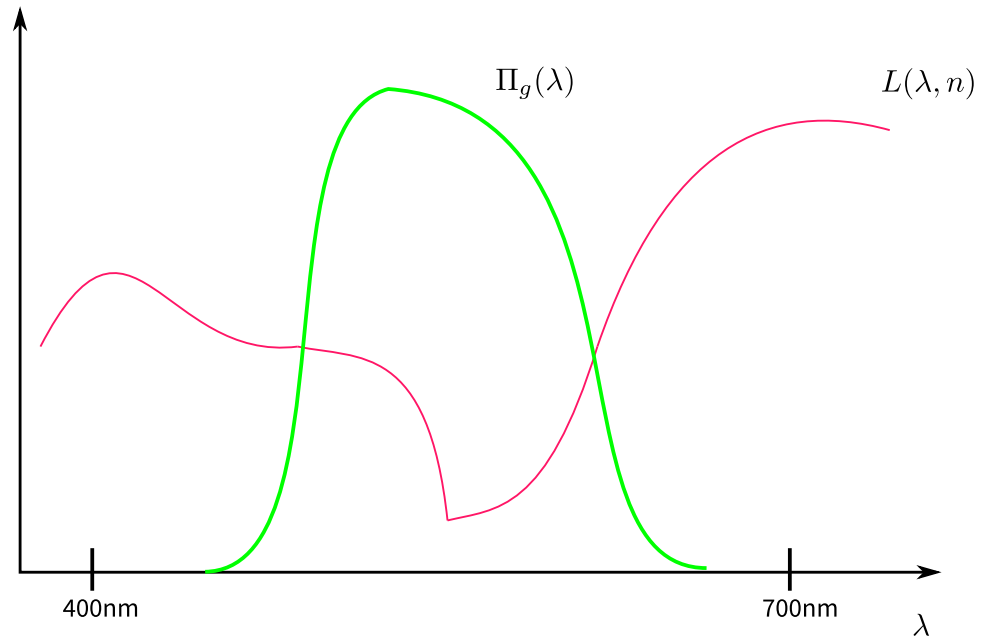
# COLOR



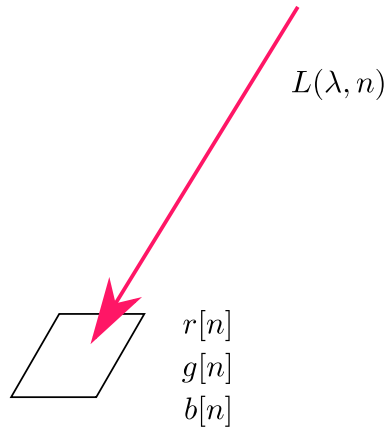
$$r[n] = \int L(\lambda, n) \Pi_r(\lambda) d\lambda$$

$$g[n] = \int L(\lambda, n) \Pi_g(\lambda) d\lambda$$

$$b[n] = \int L(\lambda, n) \Pi_b(\lambda) d\lambda$$



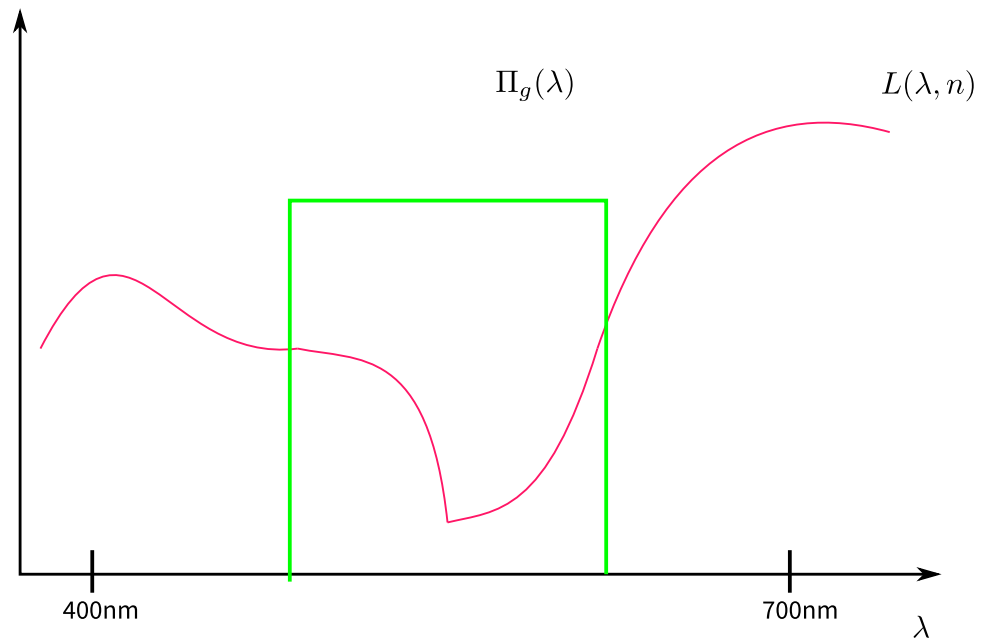
# COLOR



$$r[n] = \int L(\lambda, n) \Pi_r(\lambda) d\lambda$$

$$g[n] = \int L(\lambda, n) \Pi_g(\lambda) d\lambda$$

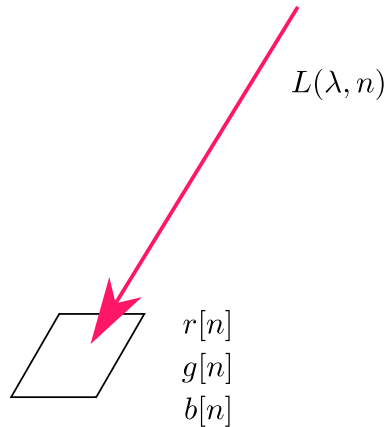
$$b[n] = \int L(\lambda, n) \Pi_b(\lambda) d\lambda$$



Simple View:

Total / Average Intensity in "Green Part" of the spectrum.

# COLOR

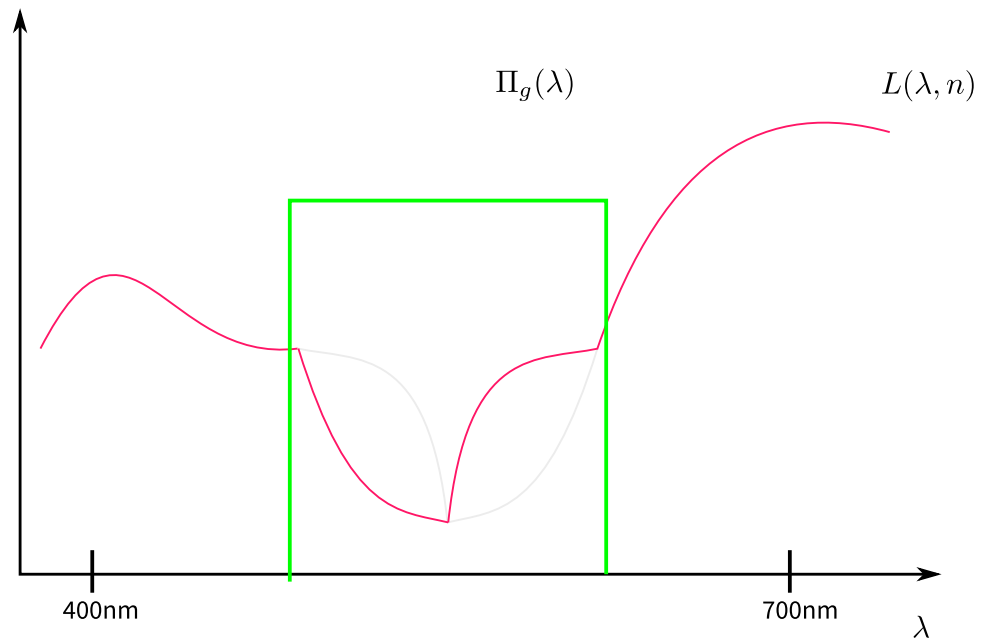


$$r[n] = \int L(\lambda, n) \Pi_r(\lambda) d\lambda$$

$$g[n] = \int L(\lambda, n) \Pi_g(\lambda) d\lambda$$

$$b[n] = \int L(\lambda, n) \Pi_b(\lambda) d\lambda$$

Metamers: Different  $L$  that have the same measured RGB values.



Simple View:

Total / Average Intensity in "Green Part" of the spectrum.

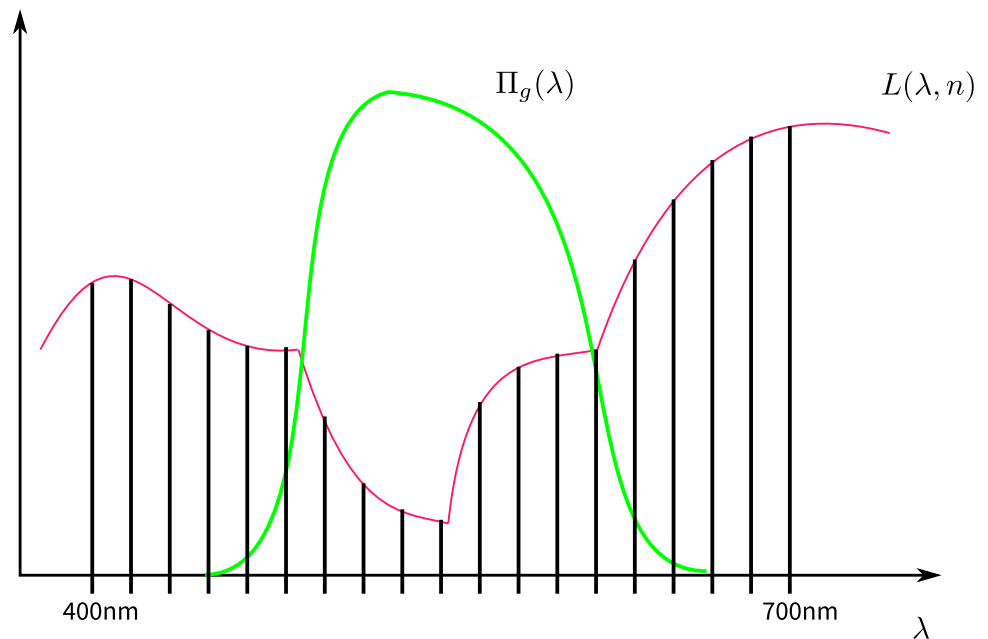
# COLOR

For simplicity,  
Have discrete wavelengths  
Approximate integration as summation

$$r[n] = \int L(\lambda, n) \Pi_r(\lambda) d\lambda$$

$$g[n] = \int L(\lambda, n) \Pi_g(\lambda) d\lambda$$

$$b[n] = \int L(\lambda, n) \Pi_b(\lambda) d\lambda$$



$$L(\lambda, n) \rightarrow L[\lambda, n] \text{ or } L[n] \in \mathbb{R}^B$$



# COLOR

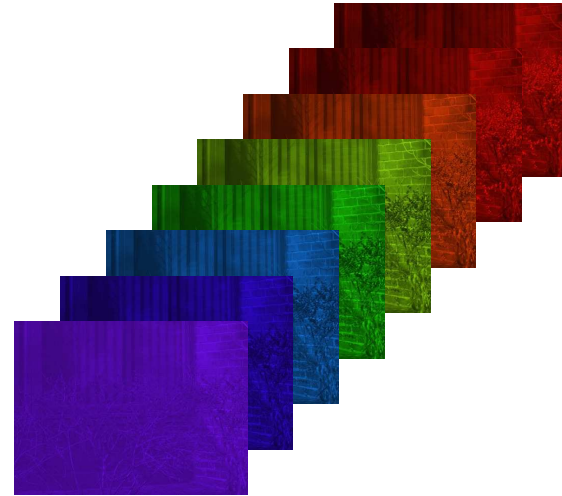
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$$r[n] = \langle L[n], \Pi_r \rangle$$

$$g[n] = \langle L[n], \Pi_g \rangle$$

$$b[n] = \langle L[n], \Pi_b \rangle$$



Think of the incident light being  
a  $B$  ( $\gg 3$ ) channel image  $L[n]$

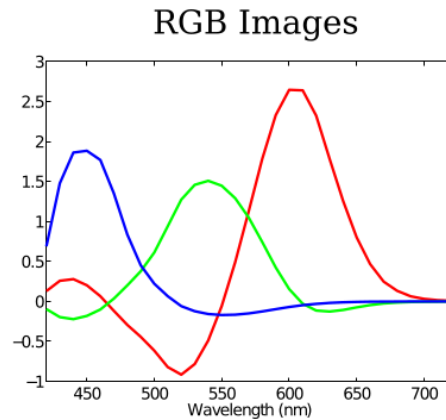
$$L(\lambda, n) \rightarrow L[\lambda, n] \text{ or } L[n] \in \mathbb{R}^B$$

# COLOR



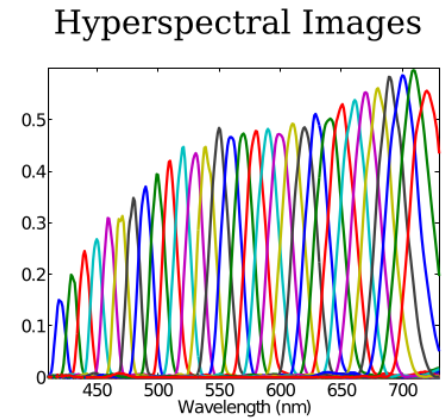
$X[n]$

I



3 Spectral Samples per pixel

VS



Higher "spectral resolution"

There are cameras that actually capture such "hyperspectral" images.

$$L(\lambda, n) \Rightarrow L[\lambda, n] \text{ or } L[n] \in \mathbb{R}^B$$

# COLOR

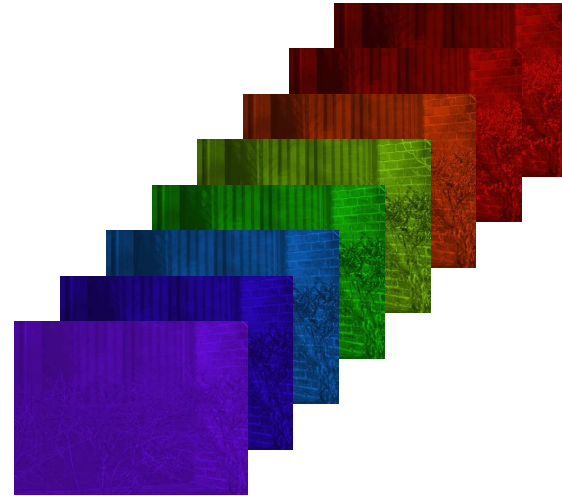
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$$X[n] = \Pi^T L[n],$$

$$\Pi = [\Pi_r \quad \Pi_g \quad \Pi_b]$$

(B x 3 Matrix)



Think of the incident light being  
a B ( $\gg 3$ ) channel image  $L[n]$

- 3 Dimensional Projection from higher  
dimensional space

- Invariant to changes in the "null space" of  $\Pi$

# COLOR

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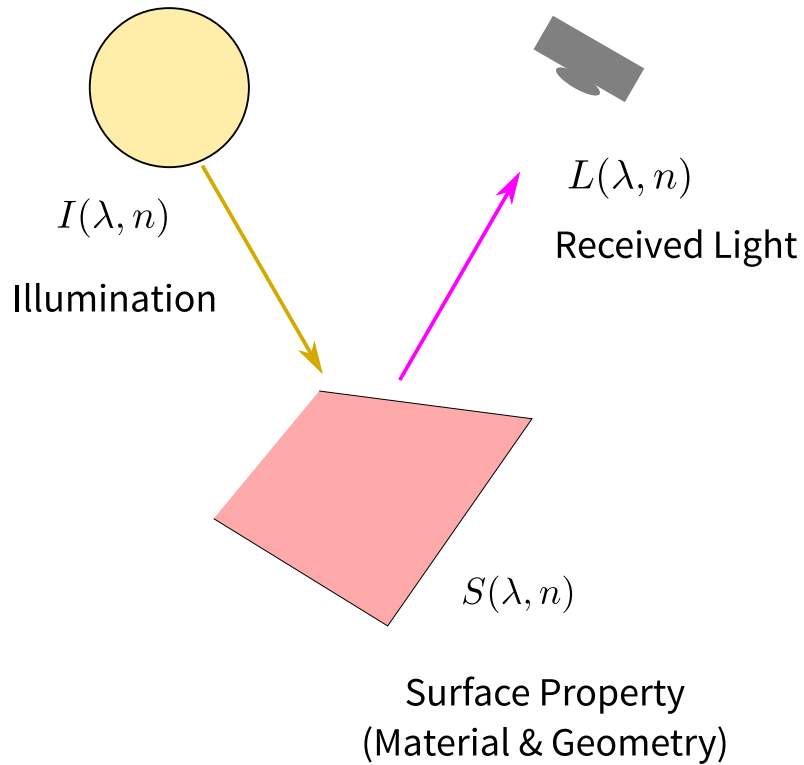
Light Color vs Object Color



# COLOR

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## Light Color vs Object Color

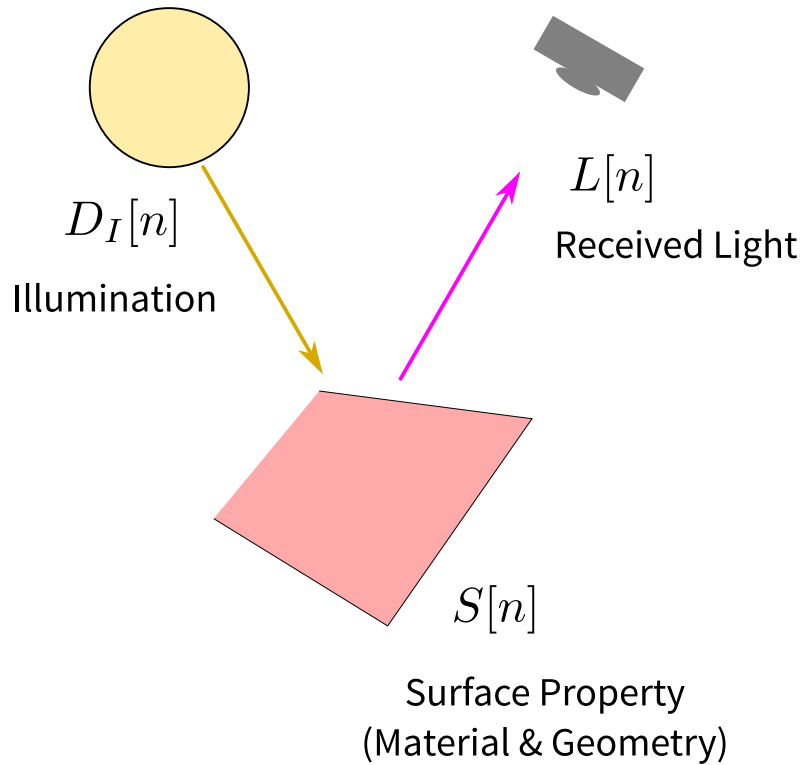


$$L(\lambda, n) = S(\lambda, n) I(\lambda, n)$$

# COLOR

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## Light Color vs Object Color

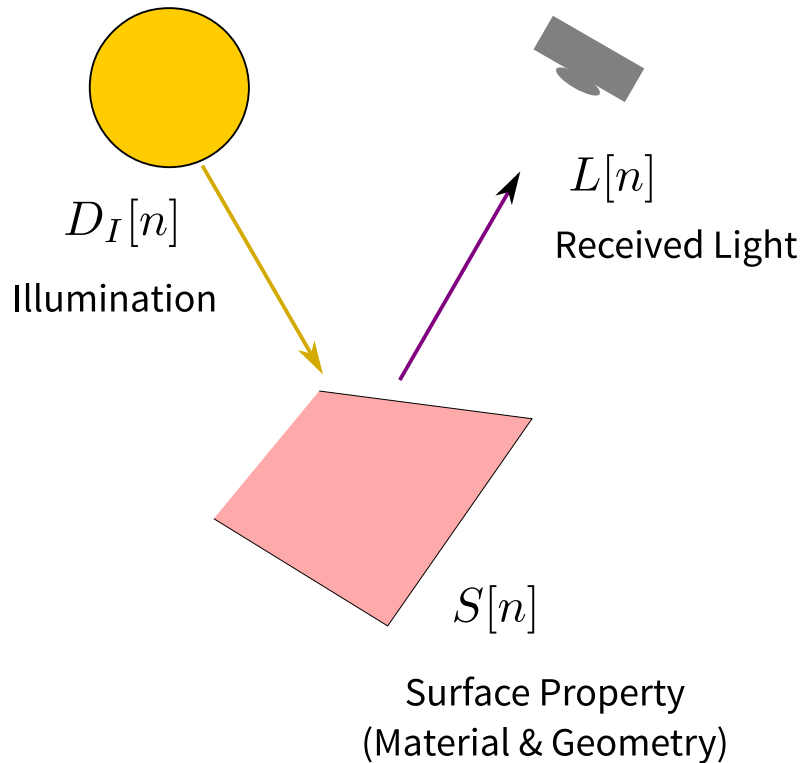


$$L[n] = D_I[n] S[n]$$

# COLOR

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## Light Color vs Object Color



$$L[n] = D_I[n] S[n]$$

Why is this important ?

Because observed color can change with illuminant color.

# COLOR

---

## Light Color vs Object Color



$$L[n] = D_I[n] S[n]$$



Why is this important ?

Because observed color can change with illuminant color.



# COLOR

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## Quick Aside: Color Temperature

Natural Illuminants often well-modeled as "black body radiators" at different "temperatures"

$$I(\lambda) \propto \frac{1}{\lambda^5} \frac{1}{\exp(\frac{hc}{k\lambda T}) - 1}$$

$h$  is Planck's constant,  $k$  is Boltzmann's constant, and  
 $T$  is "color temperature"

Warmer / cooler colors = Colors observed under illuminants with higher / lower temperature  $T$

# COLOR

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## Light Color vs Object Color



$$X_1[n] = \Pi^T (D_{I_1}[n] S[n])$$

$$L[n] = D_I[n] S[n]$$

$$X[n] = \Pi^T L[n]$$



$$X_2[n] = \Pi^T (D_{I_2}[n] S[n])$$

# COLOR

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## Light Color vs Object Color



$$X_1[n] = \Pi^T (D_{I_1}[n]S[n])$$

Even if you knew  $I_1$  and  $I_2$ ,  
could you go from  $X_1$  to  $X_2$ ?

NO



$$X_2[n] = \Pi^T (D_{I_2}[n]S[n])$$

# COLOR

## Light Color vs Object Color

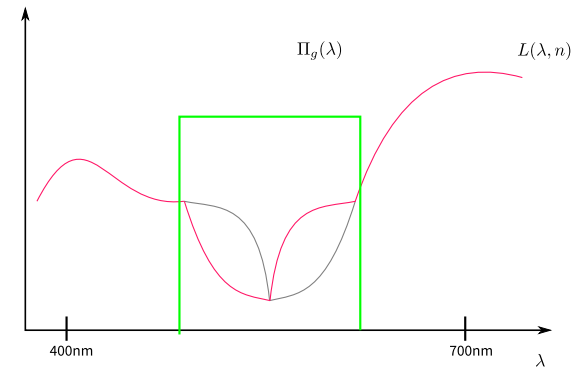


$$X_1[n] = \Pi^T (D_{I_1}[n]S[n])$$

Possible for two distinct values of  $S$  to go to same RGB value of  $X$  under  $I_1$  but **not**  $I_2$ .



$$X_2[n] = \Pi^T (D_{I_2}[n]S[n])$$



# COLOR

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## Light Color vs Object Color



Chong, Gortler, and Zickler, "The von Kries Hypothesis and a Basis for Color Constancy", ICCV 2007

$$X_1[n] = \Pi^T (D_{I_1}[n]S[n])$$

Approximations:

- Ver 1:  $X_1[n] = C_{2 \rightarrow 1} X_2[n]$

Linear Transform

$$X_1[n] = C D_{2 \rightarrow 1} C^T X_2[n]$$

Diagonal Transform  
in some color space  
(doesn't depend on 1,2)

- Ver 2:  $X_1[n] = D_{2 \rightarrow 1} X_2[n]$

Diagonal Transform  
"von-Kries" model

$$X_2[n] = \Pi^T (D_{I_2}[n]S[n])$$

# COLOR

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## Light Color vs Object Color



$$X_1[n] = \Pi^T (D_{I_1}[n]S[n])$$

$$X_1[n] = D_{2 \rightarrow 1} X_2[n]$$

But this is only if we know the illuminants  $I_1$  and  $I_2$ .



Looking at an image and separating illuminant and surface colors: color constancy

$$X_2[n] = \Pi^T (D_{I_2}[n]S[n])$$



# COLOR

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## Color Constancy



$$X_1[n] = \Pi^T (D_{I_1} \quad S[n])$$

- Assume single (or dominant) illuminant for the scene
- Say illuminant  $I_1$  is some known "canonical illuminant"
- Our task is, given  $X_2[n]$  under unknown illumination  $I_2$ , to go to an image of the scene under the canonical illuminant.

$$X_1[n] = \underline{D_{2 \rightarrow 1}} X_2[n]$$

Estimate  
Transform

$$X_2[n] = \Pi^T (D_{I_2} \quad S[n])$$

- Color constancy is about removing color cast, not say 'brightness'.
- Need to estimate transform upto scale, e.g., assume elements add upto 3

# COLOR

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## Color Constancy



$$X_1[n] = \begin{bmatrix} d_r & & \\ & d_g & \\ & & d_b \end{bmatrix} X_2[n]$$

- Have a prior on natural colors (under canonical illumination)
- Find transform so that colors in  $X_1$  match the prior



### Prior 1: Gray World

- On average, colors are neutral or gray

$$d_r \propto \frac{1}{\text{Mean}_n(X_2^r[n])}$$

Fails if there's a dominant color in the scene (like lots of grass)



# COLOR

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## Color Constancy



$$X_1[n] = \begin{bmatrix} d_r & & \\ & d_g & \\ & & d_b \end{bmatrix} X_2[n]$$

- Have a prior on natural colors (under canonical illumination)
- Find transform so that colors in  $X_1$  match the prior



## Prior 2: White Patch Retinex

- The brightest color is neutral (or white)

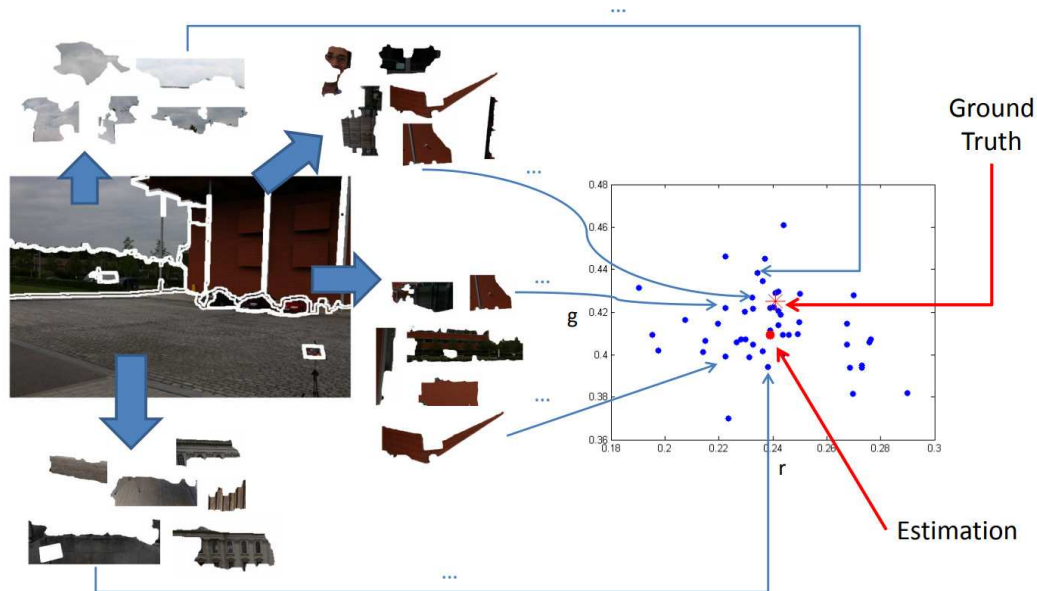
$$d_r \propto \frac{1}{\text{Max}_n(X_2^r[n])}$$

Probably want to average over the highest  $k$  values instead of just max.

# COLOR

## Color Constancy

- Remains an active area of research. Modern methods include using CNNs, trying to match objects, and so on.



Joze & Drew, Exemplar-based Color Constancy and Multiple Illumination, PAMI 2014

## Color Constancy

- Remains an active area of research. Modern methods include using CNNs, trying to match objects, and so on.
- Humans solve it well, but vision scientists still trying to figure out how



# COLOR

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## Color Constancy

- Remains an active area of research. Modern methods include using CNNs, trying to match objects, and so on.
- Humans solve it well, but vision scientists still trying to figure out how
- Multi-illuminant case even more interesting. Efforts to solve it with more information.



(a) No-flash / flash images

(b) Lighting separation

(c) White-balance

(d) Light editing

Hui et al., Post-Capture Lighting Manipulation using Flash Photography, arXiv 1704.05564

# COLOR

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## Color Representation / Spaces

- We're using 3 numbers to represent color: co-ordinates in some space.
- But RGB may not always be the right way to "work" with color.
- The right representation depends on what you're trying to do:
  - Analyze an Image
  - Restore a degraded Image
  - Manipulate an Image for Visual / Artistic Quality
  - Print / Display Images
  - Measure similarity between two images

# COLOR

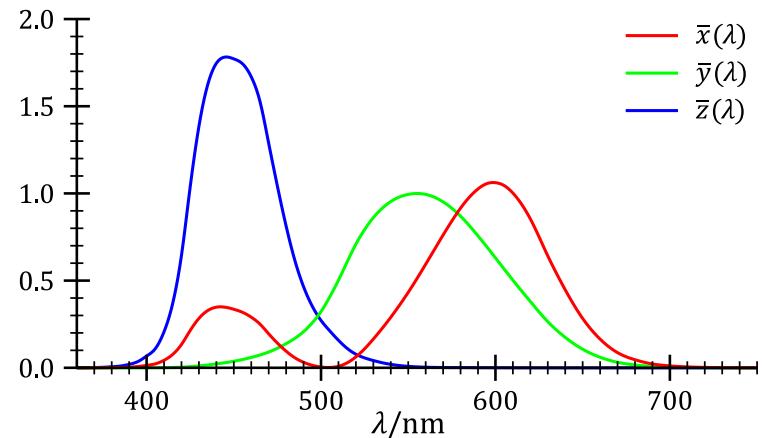
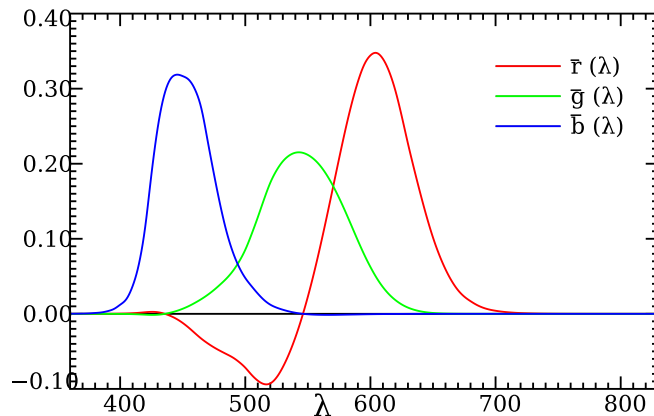
## Color Representation / Spaces

Linear Spaces: Some linear transformation of RGB

$$X_{ABC} = T \ X_{RGB}$$

$$\begin{bmatrix} R \\ G \\ B \end{bmatrix} = \begin{bmatrix} 0.418,47 & -0.158,66 & -0.082,835 \\ -0.091,169 & 0.252,43 & 0.015,708 \\ 0.000,920,90 & -0.002,549,8 & 0.178,60 \end{bmatrix} \cdot \begin{bmatrix} X \\ Y \\ Z \end{bmatrix}$$

- XYZ: This is a CIE standard, because the RGB filter functions are actually negative !



Source: Wikipedia

# COLOR

## Color Representation / Spaces

Linear Spaces: Some linear transformation of RGB

$$X_{ABC} = T \ X_{RGB}$$

- YUV: Useful as a "decorrelating" transform.

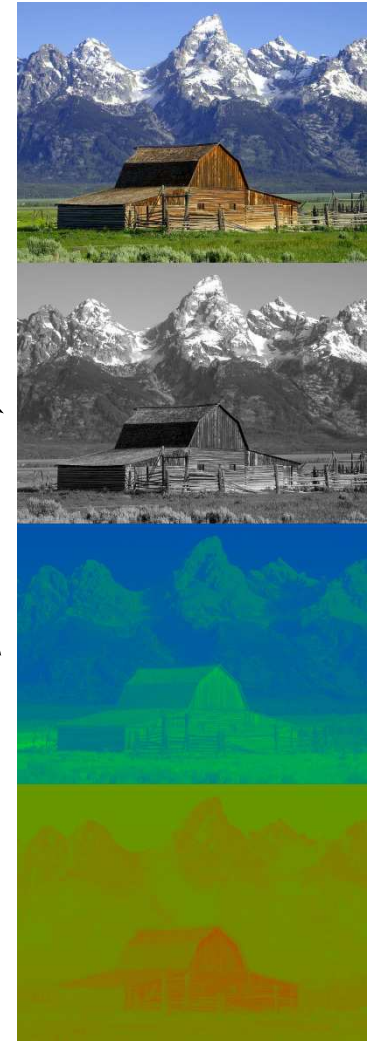
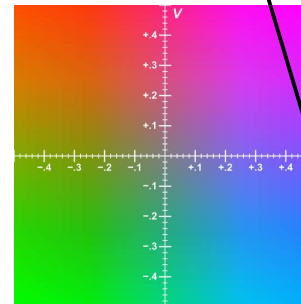
$$\begin{bmatrix} Y' \\ U \\ V \end{bmatrix} = \begin{bmatrix} 0.299 & 0.587 & 0.114 \\ -0.14713 & -0.28886 & 0.436 \\ 0.615 & -0.51499 & -0.10001 \end{bmatrix} \begin{bmatrix} R \\ G \\ B \end{bmatrix},$$
$$\begin{bmatrix} R \\ G \\ B \end{bmatrix} = \begin{bmatrix} 1 & 0 & 1.13983 \\ 1 & -0.39465 & -0.58060 \\ 1 & 2.03211 & 0 \end{bmatrix} \begin{bmatrix} Y' \\ U \\ V \end{bmatrix}.$$

Think of Y as grayscale image.

Called luma, luminance, lightness

U and V is the "color information"

May be useful to  
process these  
independently  
(instead of R,G,B)



# COLOR

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## Color Representation / Spaces

### Non-Linear Transforms

#### Lightness and rg Chromaticity

$$L = (R + G + B), r = R/L, g = G/L$$

Could also do this in XYZ (xy-chromaticity).

Main idea is that we're representing colors as ratios, instead of linear transform.

Chromaticity values stay constant with changing brightness, UV values don't.



Input



Y / L



Histogram  
Equalized



Keep U,V from  
original image



Keep r,g from  
original image



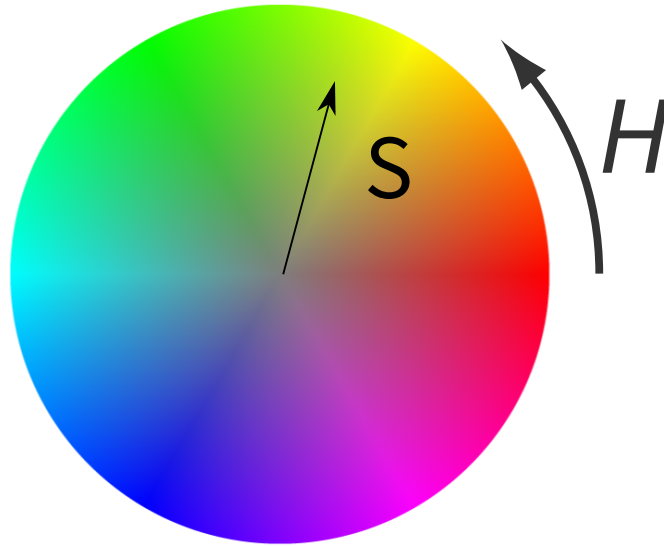
# COLOR

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Color Representation / Spaces

Non-Linear Transforms

Other ways of representing chromaticity: Hue/Saturation



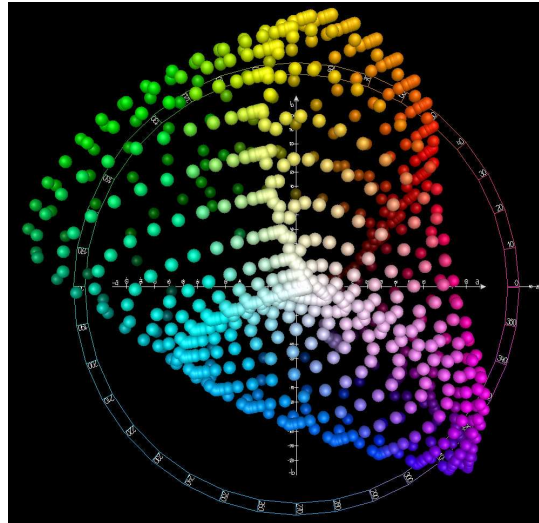
# COLOR

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## Color Representation / Spaces

### Non-Linear Transforms

CIE LAB color space: Distances in Lab are "perceptually meaningful"

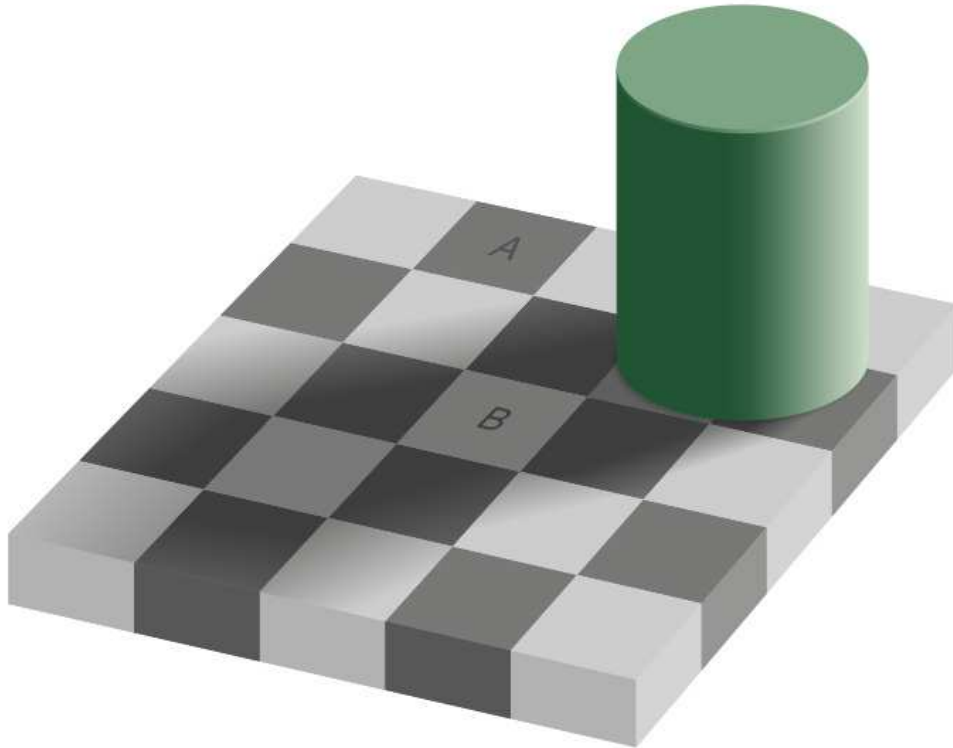


Read: Wikipedia Article on CIE-LAB for exact definition.

# SHADING

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Observed color isn't the only thing that changes with the environment. So does lightness.



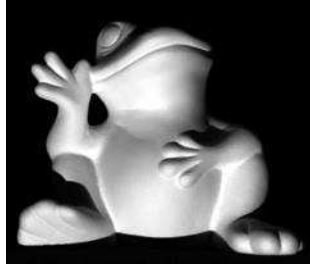
Squares A and B have exactly the same intensity.

But we don't just want brightness constancy !

Source: Ted Edelson

# SHADING

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The way intensity changes over a constant surface, in a single image, or with change in lighting, can be a strong cue for shape.

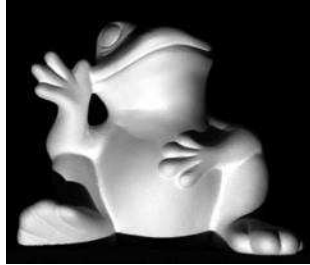


Sculptors have used it to fake depth.

Source: Belhumeur, Kriegman, and Yuille

# SHADING

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The way intensity changes over a constant surface, in a single image, or with change in lighting, can be a strong cue for shape.



Source: Debevec

Computer Vision has  
used it for shape capture

# SHADING

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Shading can be quite complex, depending on the material !